

Evaluating the Potential of a Blocking Predictor in a Hybridized Dynamical-Statistical Model for Improved Week 3-4 Temperature and Precipitation Outlooks

Cory Baggett

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Innovim, LLC / NOAA / NWS / CPC

Week 3-4 / S2S Webinar presented by **NOAA OAR/WPO** and **NWS/OSTI** on

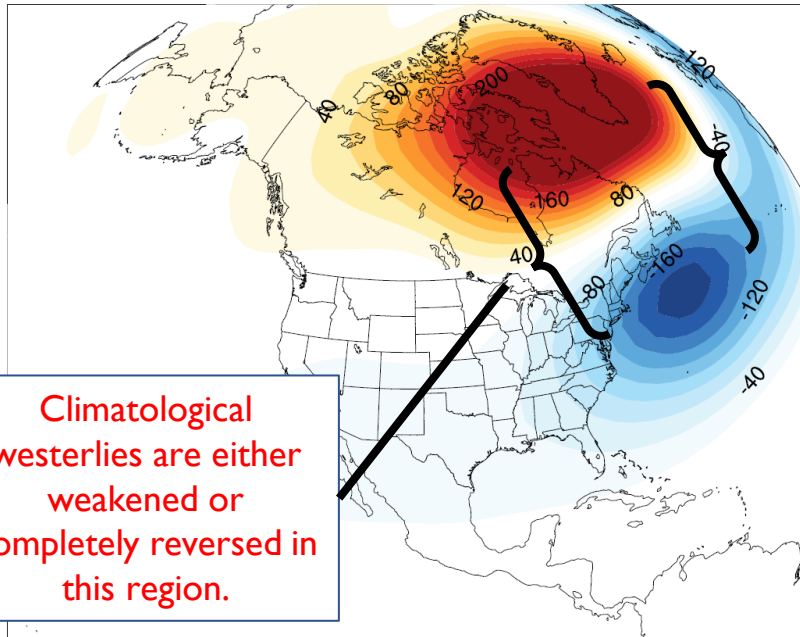
September 13, 2021

Work funded by the **Modeling Program Division, Office of Science and Technology Integration, NWS**

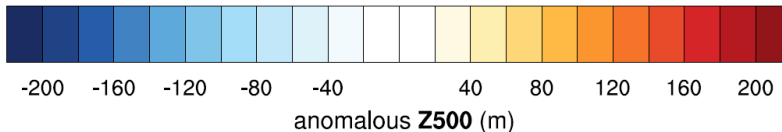
Why do we care about blocking?

Composite of Anomalous Z500

(300E blocked during DJF)



Climatological westerlies are either weakened or completely reversed in this region.



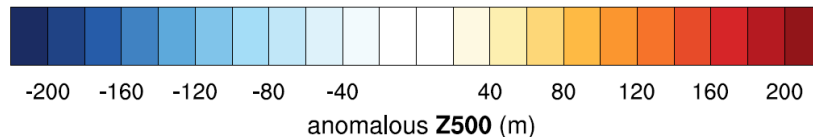
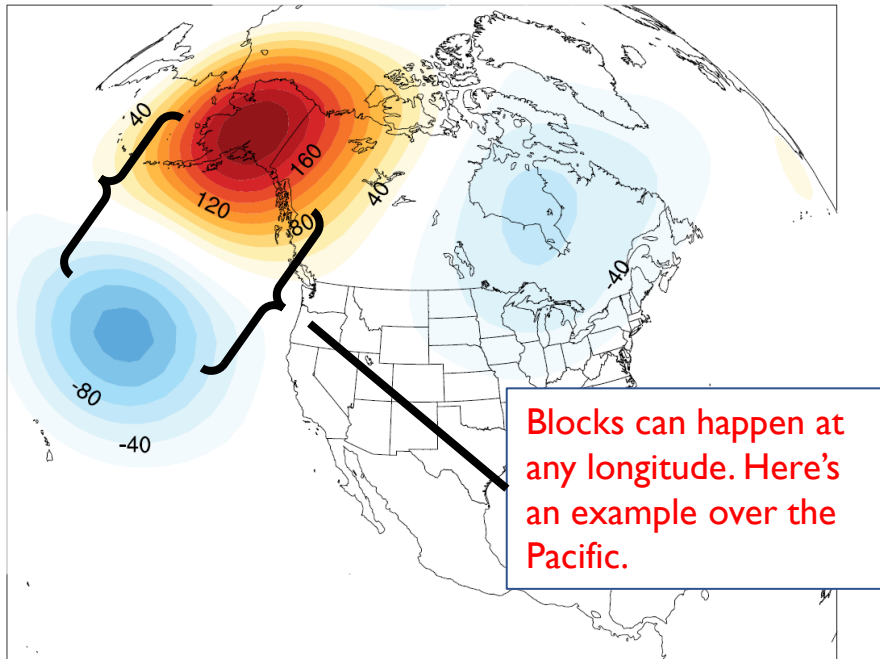
During blocking:

- An anomalous ridge exists to the north and an anomalous trough to its south.
- This results in a reversal of the climatological westerlies to easterlies.
- This reversal blocks the jet stream, forcing large-scale stationary waves and a diversion of the storm track.
- This pattern resembles the negative phase of the North Atlantic Oscillation (NAO).

The Tibaldi and Molteni (1990) index is one amongst many blocking indices (Barnes et al. 2012).

Composite of Anomalous Z500

(210E blocked during JFM)



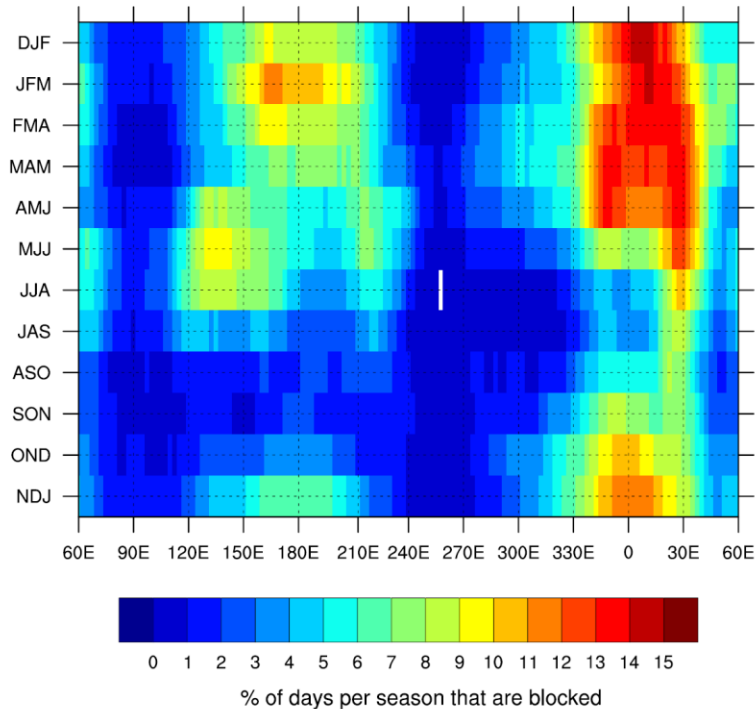
Examples of impacts:

- Extensive drought in the West (Wise 2016)
- Divert atmospheric rivers into Alaska (Baggett et al. 2015)
- **Extreme cold conditions** (Wang et al. 2010; Marinaro et al. 2015)
- Sudden stratospheric warmings (Martius et al. 2009; Butler et al. 2017)

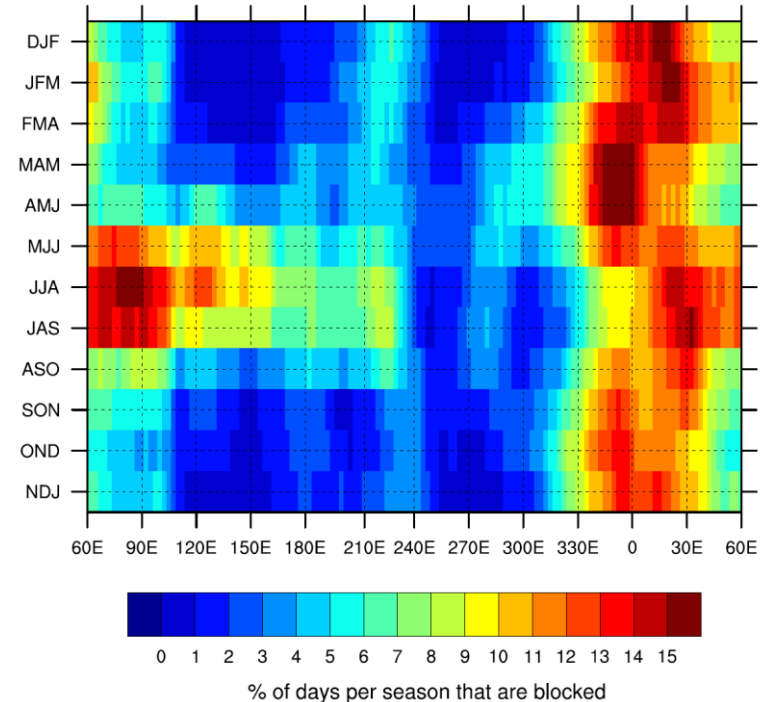
Because blocks can persist for weeks, knowledge of blocking episodes and their surface impacts can perhaps lead to enhanced predictive skill of Week 3-4 temperature and precipitation across the United States.

Climatological Blocking Frequency

TM1990 Blocking Climatology (1979-2020)



B2012 Blocking Climatology (1979-2020)



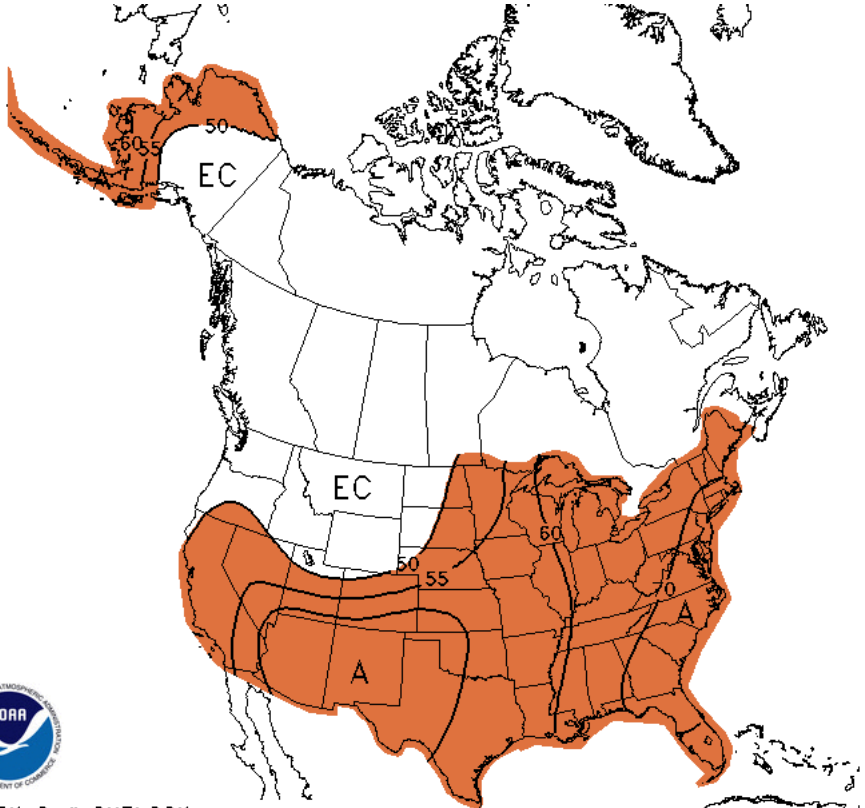
- Blocking occurs most frequently over the Atlantic sector.
- Which index should we use? Which blocking longitude?
- We tried many blocking-related indices, but we have found using the North Atlantic Oscillation (NAO) for the Atlantic and the Pacific-North American pattern (PNA) for the Pacific as “blocking” indices work well (Croci-Maspoli et al. 2007).
- Forthcoming results shown in this presentation use the NAO and PNA.

A brief overview of CPC's forecasting process...

CPC Forecast Tools for Week 3-4



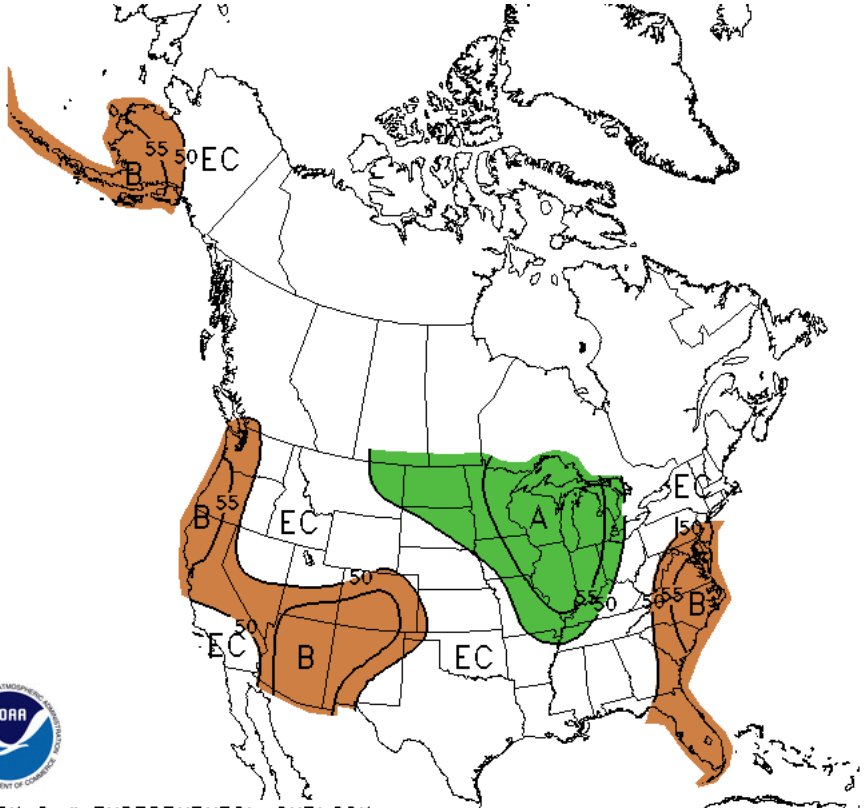
Temperature



WEEK 3-4 OUTLOOK
 TEMPERATURE PROBABILITY
 MADE 10 SEP 2021
 VALID SEP 25 - OCT 08, 2021

EC MEANS 50/50 CHANCES
 FOR ABOVE OR BELOW
 A MEANS ABOVE NORMAL
 B MEANS BELOW NORMAL

Precipitation



WEEK 3-4 EXPERIMENTAL OUTLOOK
 PRECIPITATION PROBABILITY
 MADE 10 SEP 2021
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CPC Forecast Tools for Week 3-4



- Week 3-4 outlooks are issued once per week on Friday.
- A single forecaster (rotated weekly) is assigned to make the official outlook.
- The forecaster receives input from several tools:

1. Dynamical Models

2. Statistical Models

3. Model Blends

4. Intuition and Consistency

GEFSv12, ECMWF, CFSv2, GEMv3, JMA, and models participating in SubX



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Multiple Linear Regression (MLR)*,
Phase Model, Constructed Analog,
Linear Inverse Model

*Because the dynamical models are typically deficient at simulating blocking, along with stratosphere-troposphere interactions at extended leads (Domeisen et al. 2020a,b; Quinting and Vitart 2019), we focus on improving CPC's MLR statistical model via a hybridization approach with blocking-related indices as forecasted by the dynamical models (e.g., Kim et al. 2021)

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Equal-Weighted, Manual Blend,
Autoblend



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Peer input, recent model errors,
persistence, soil moisture, coherent
MJO



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4. Intuition and Consistency

Categorical above or below normal outlooks of temperature and precipitation, averaged over Week 3-4.

- The forecaster presents a preliminary outlook at a weekly forecast discussion. Input is received, adjustments are made, and the outlook is made.
- Temperature outlooks are operational while precipitation outlooks are still experimental due to lack of skill.



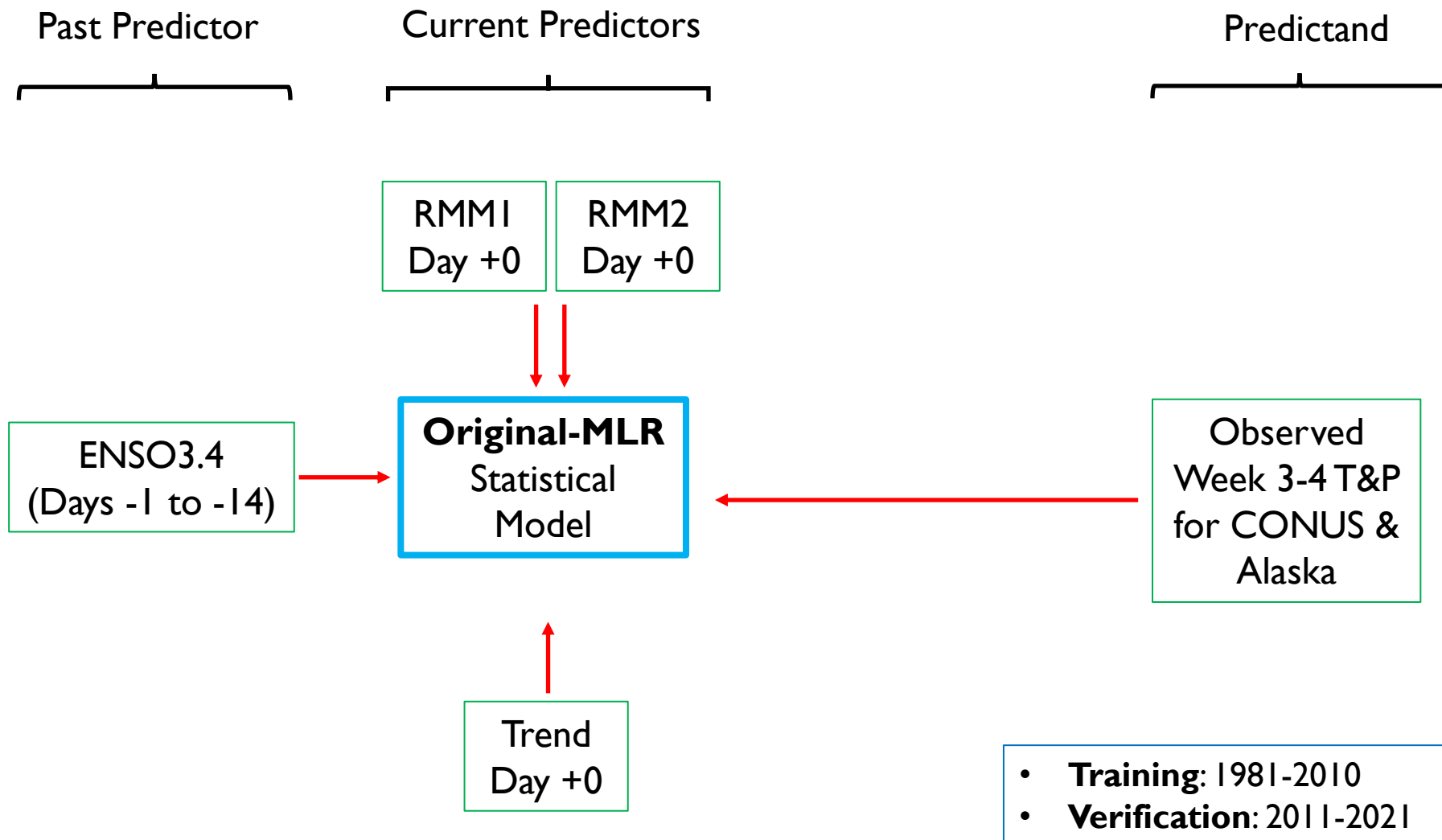
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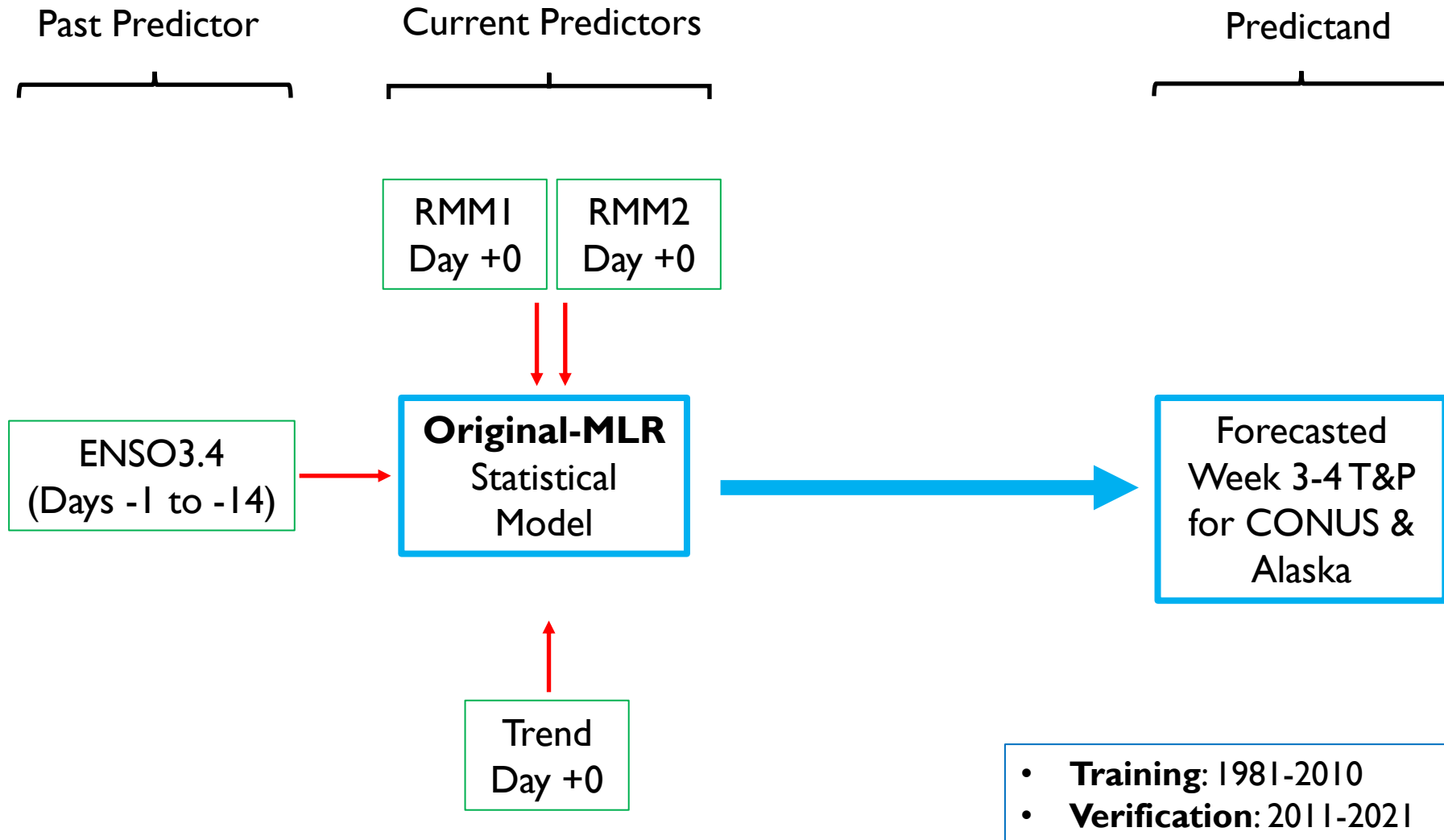
What is CPC's Multiple Linear Regression Model (original-MLR)?

Original-MLR Schematic (Training)



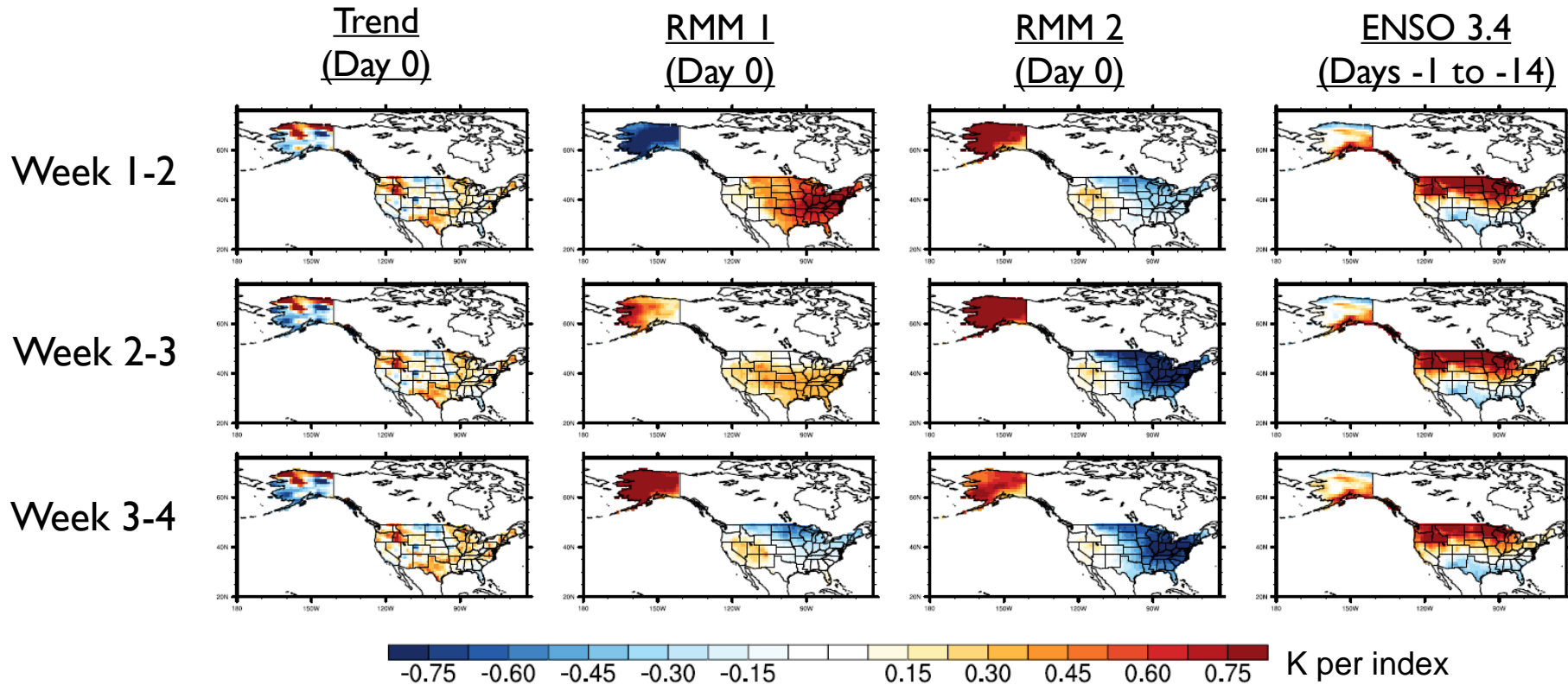
Harnos et al., in prep.

Original-MLR Schematic (Verification)



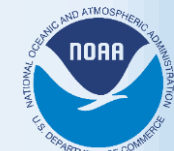
Harnos et al., in prep.

Temperature



- Trend remains constant with lead-time
- MJO and ENSO signals persist out to Week 3-4

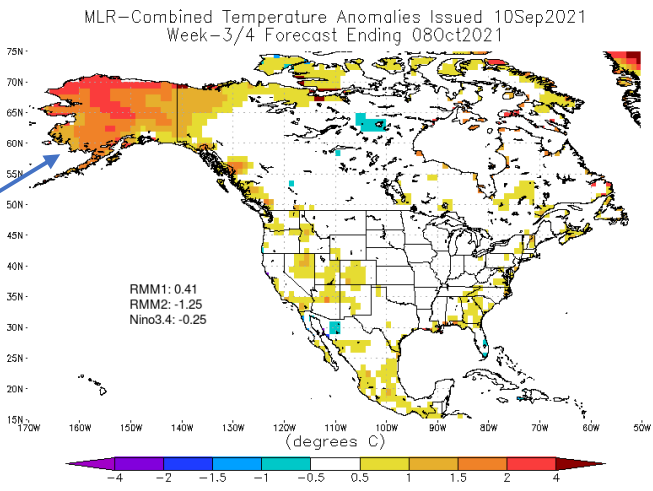
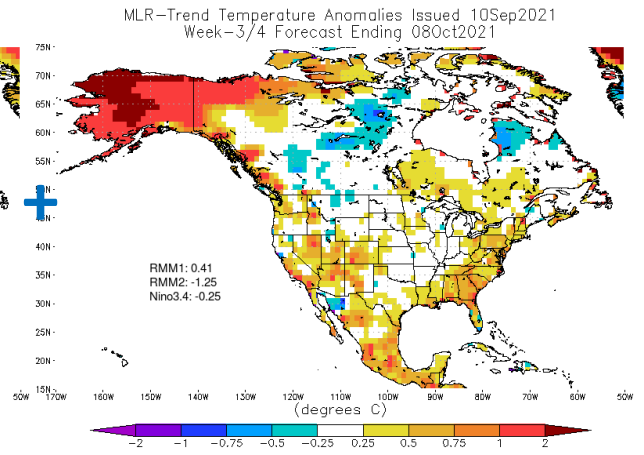
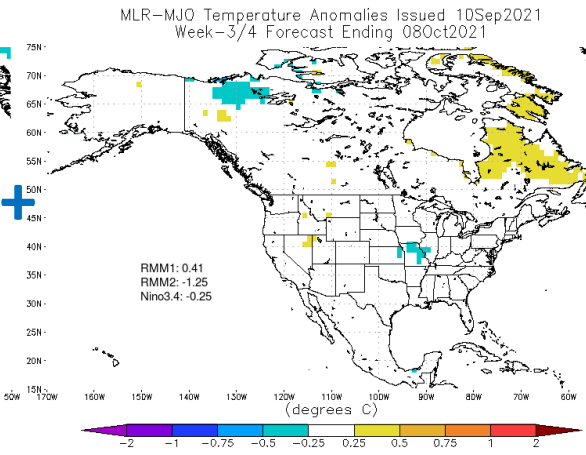
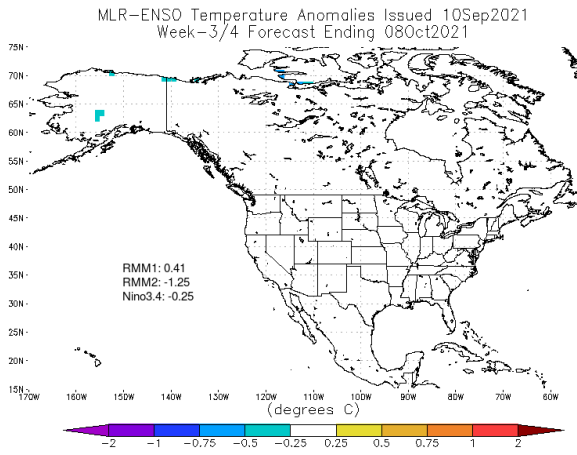
Original Multiple-Linear Regression Forecast



ENSO 3.4 (Days -1 to -14)

RMMs 1 & 2 (Day 0)

Trend (Day 0)



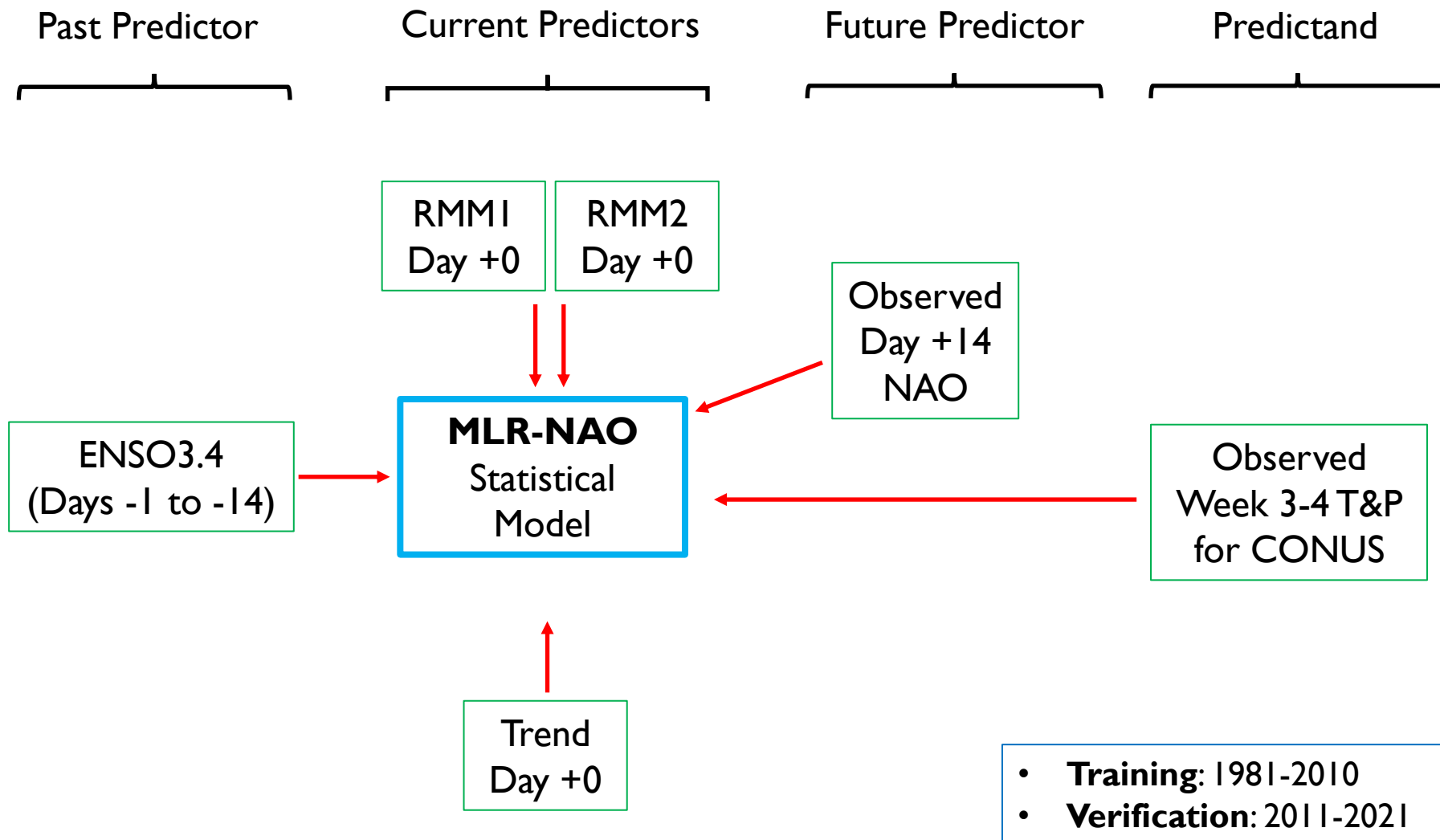
Significant warm signal in Alaska from the trend, which conflicted with some dynamical model guidance.

Week ENSO and MJO contributions, led to the trend dominating the original-MLR forecast.

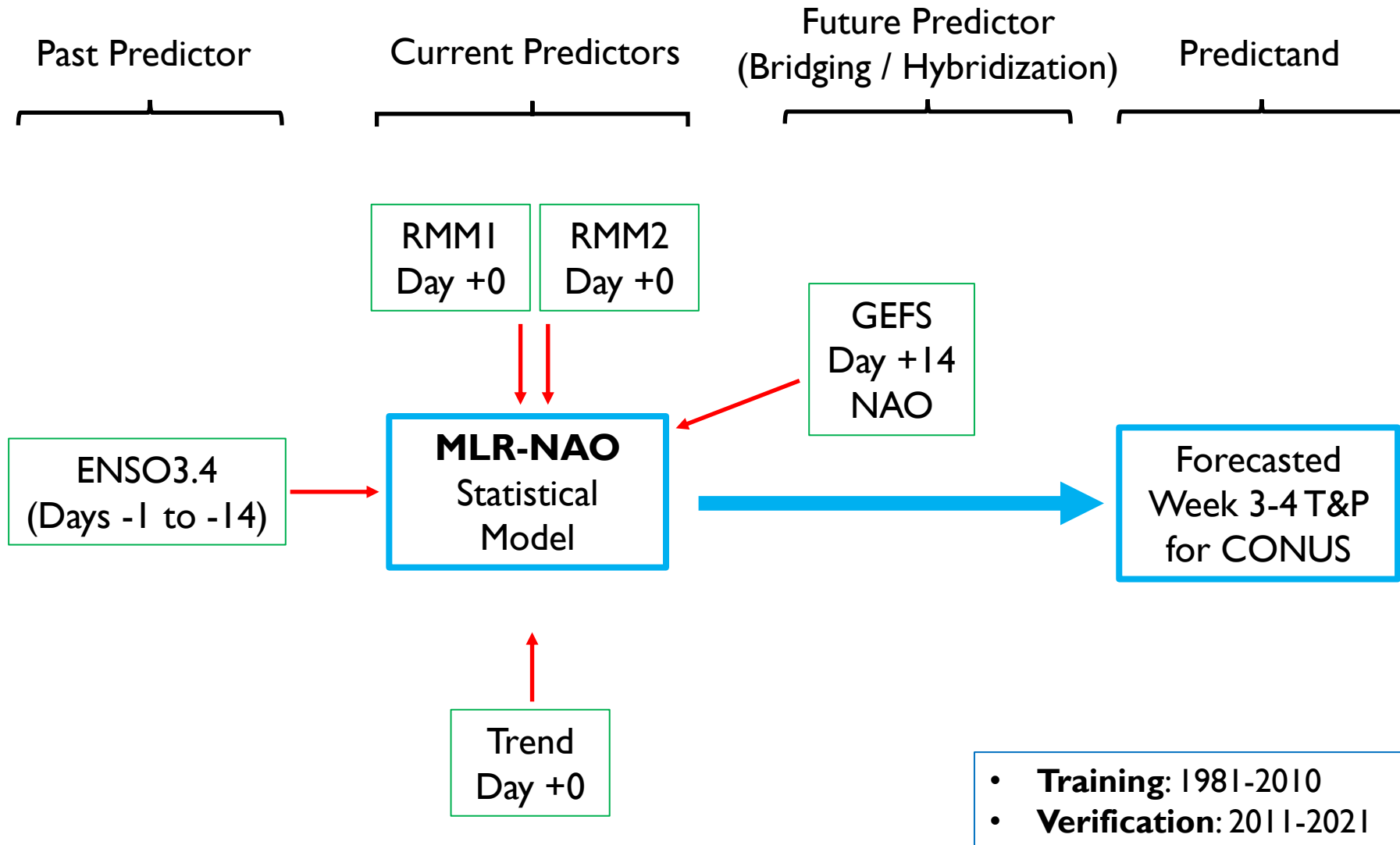
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How does an MLR-NAO perform compared to the original-MLR?

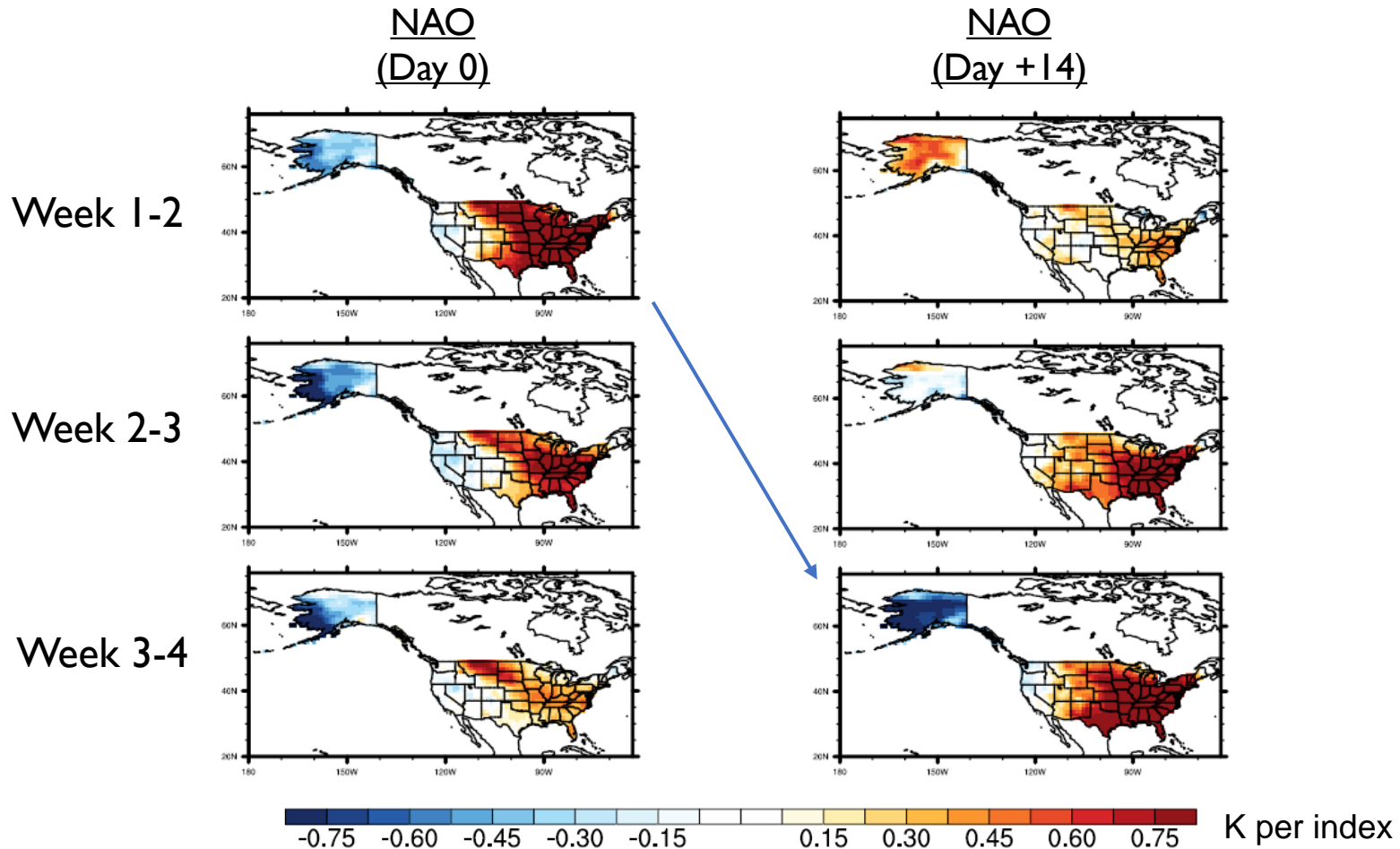
MLR-NAO Schematic (Training)



MLR-NAO Schematic (Verification)

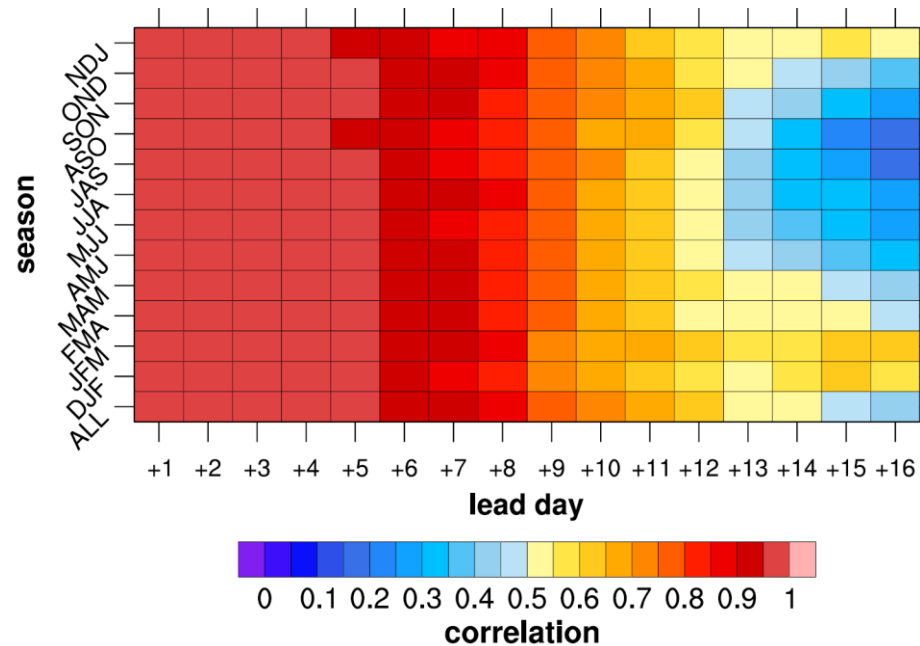


MLR-NAO Regression Coefficients



- Training the MLR-NAO on the observed Day +14 NAO provides a much stronger Week 3-4 signal, so hybridization/bridging the statistical MLR with values of the NAO forecasted by the dynamical models may be ideal.

Observed NAO versus GEFS NAO



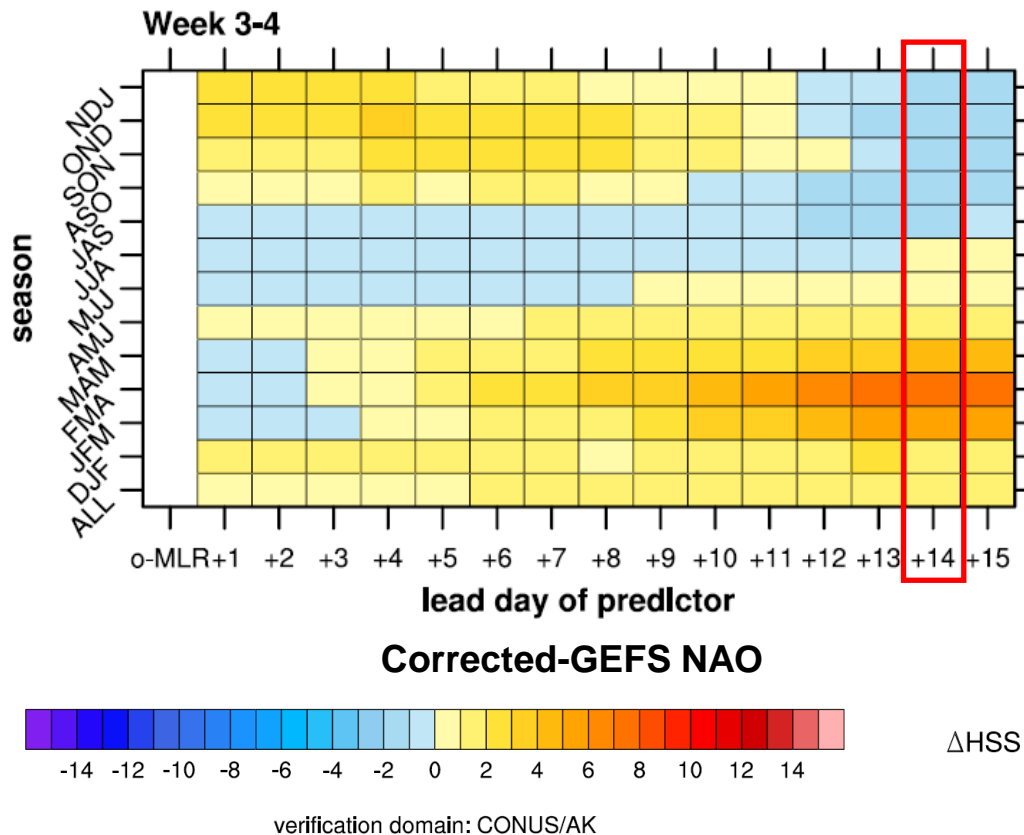
- GEFS forecast versus observations
- The NAO is based on CPC's methods (RPCA on Z500; Barnston & Livezey 1987)

• NAO skill scores peak in winter, with correlations exceeding 0.5 through Day +14, when averaged across all seasons.

Original-MLR versus MLR-NAO



Seasonal Skill Score Improvement of the MLR-NAO over the original-MLR



Difference in Week 3-4 TEMPERATURE skill scores

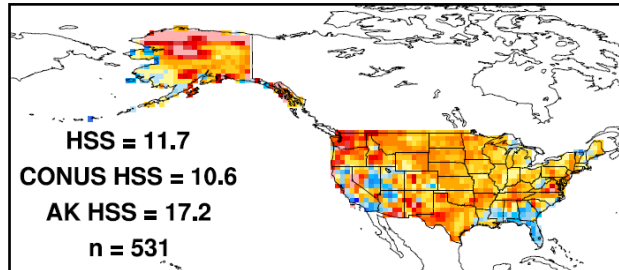
- original-MLR versus MLR-NAO
- **Predictor:** GEFS Days +1 to +15 NAO
- **Verification Period:** 2011-2021, Tuesday and Fridays
- **Key Points:**
 - Generally, the MLR-NAO offers the most improvement during non-summer months.
 - The greatest improvement is provided using the GEFS NAO from Day +14
 - One could make the argument that we should use predictors with smaller leads during fall.

original-MLR versus MLR-NAO

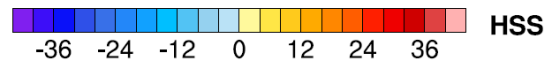
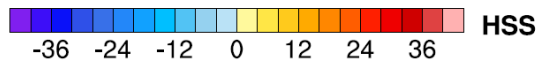
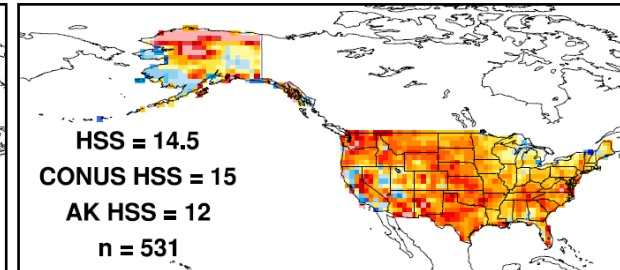
Temperature

original-MLR versus MLR-NAO during NDJFMA (2011-2021)
during All Forecast Initializations

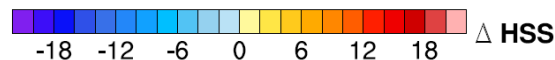
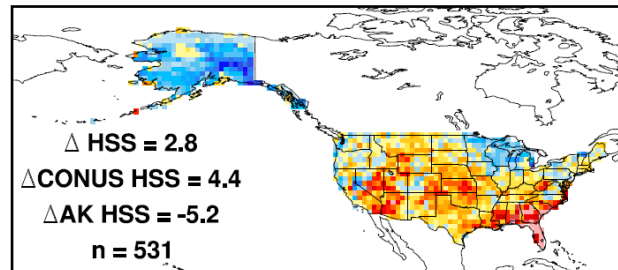
original-MLR



MLR-NAO



MLR-NAO minus original-MLR



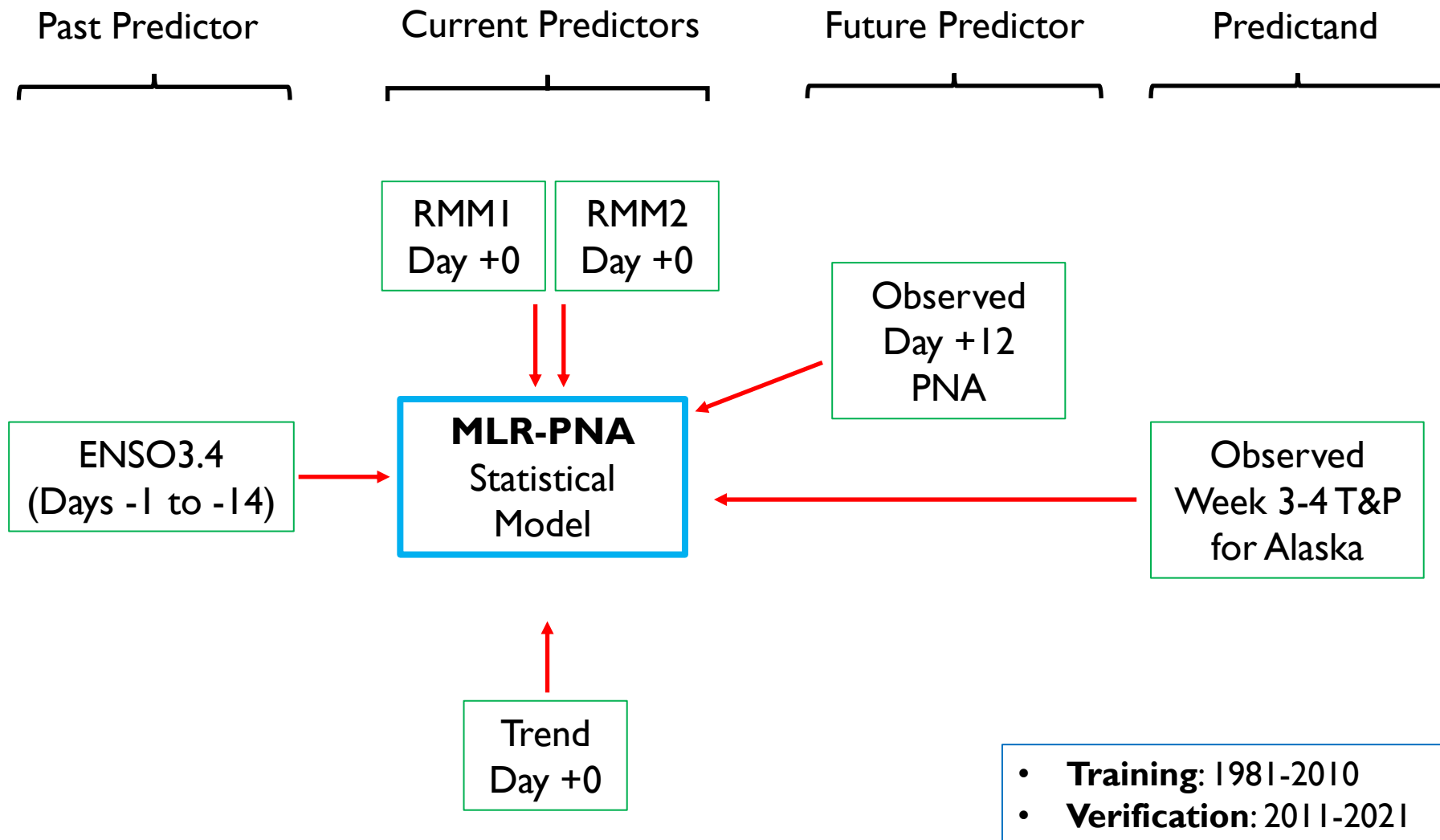
Predictor: corrected-GEFS Day +14 NAO (CPC)

Difference in Week 3-4 TEMPERATURE skill scores

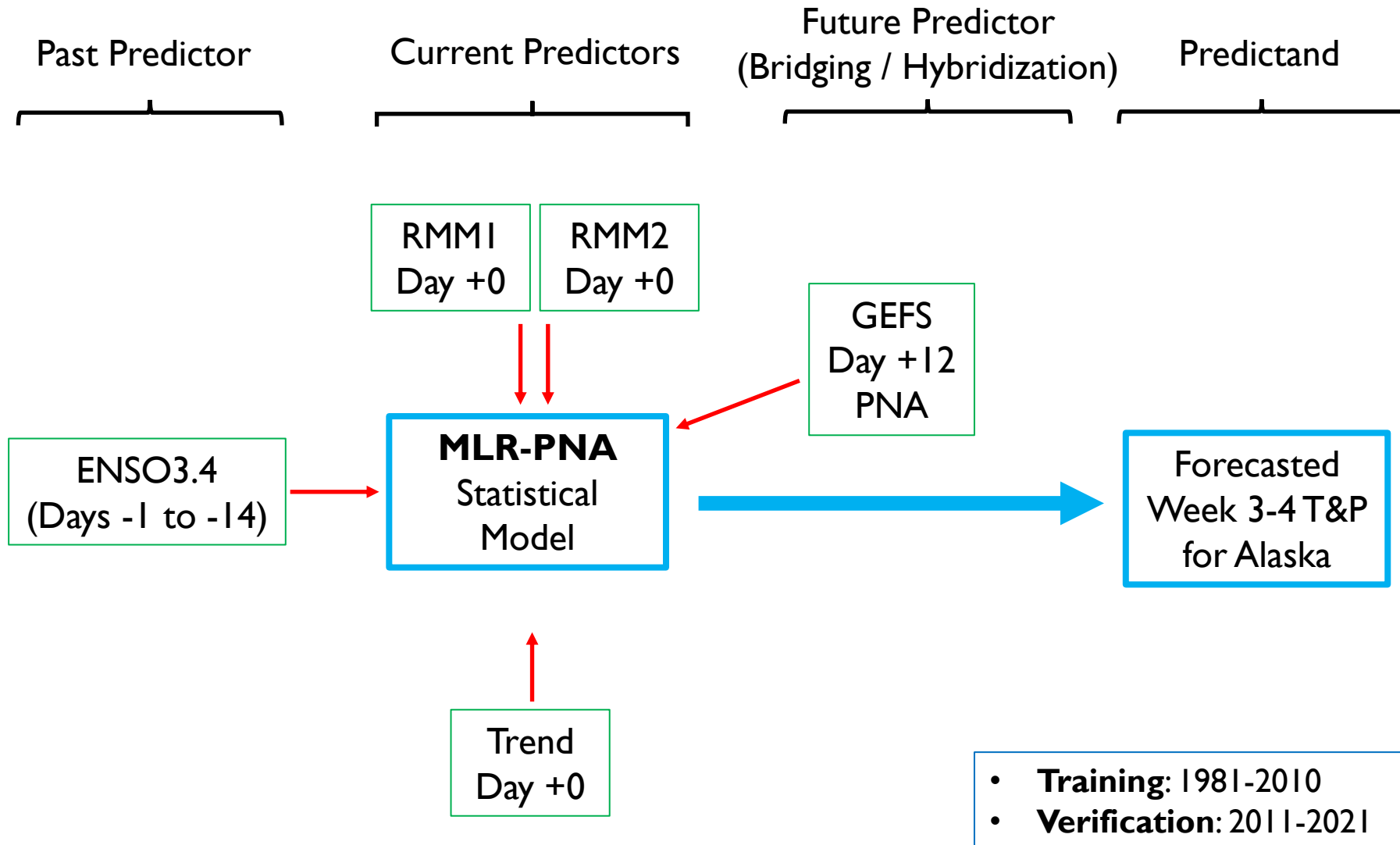
- **original-MLR versus MLR-NAO**
- **Predictor:** GEFS Day +14 NAO
- **Verification Period:** 2011-2021, Tuesday and Fridays
- **Additional Conditions:** during November-April only
- **Key Points:**
 - Overall skill scores improve by ~24%.
 - Generally, the MLR-NAO offers improvements over CONUS and makes things worse over AK.
 - Skill scores over CONUS improve by ~42%.

How does an MLR-PNA perform compared to the original-MLR?

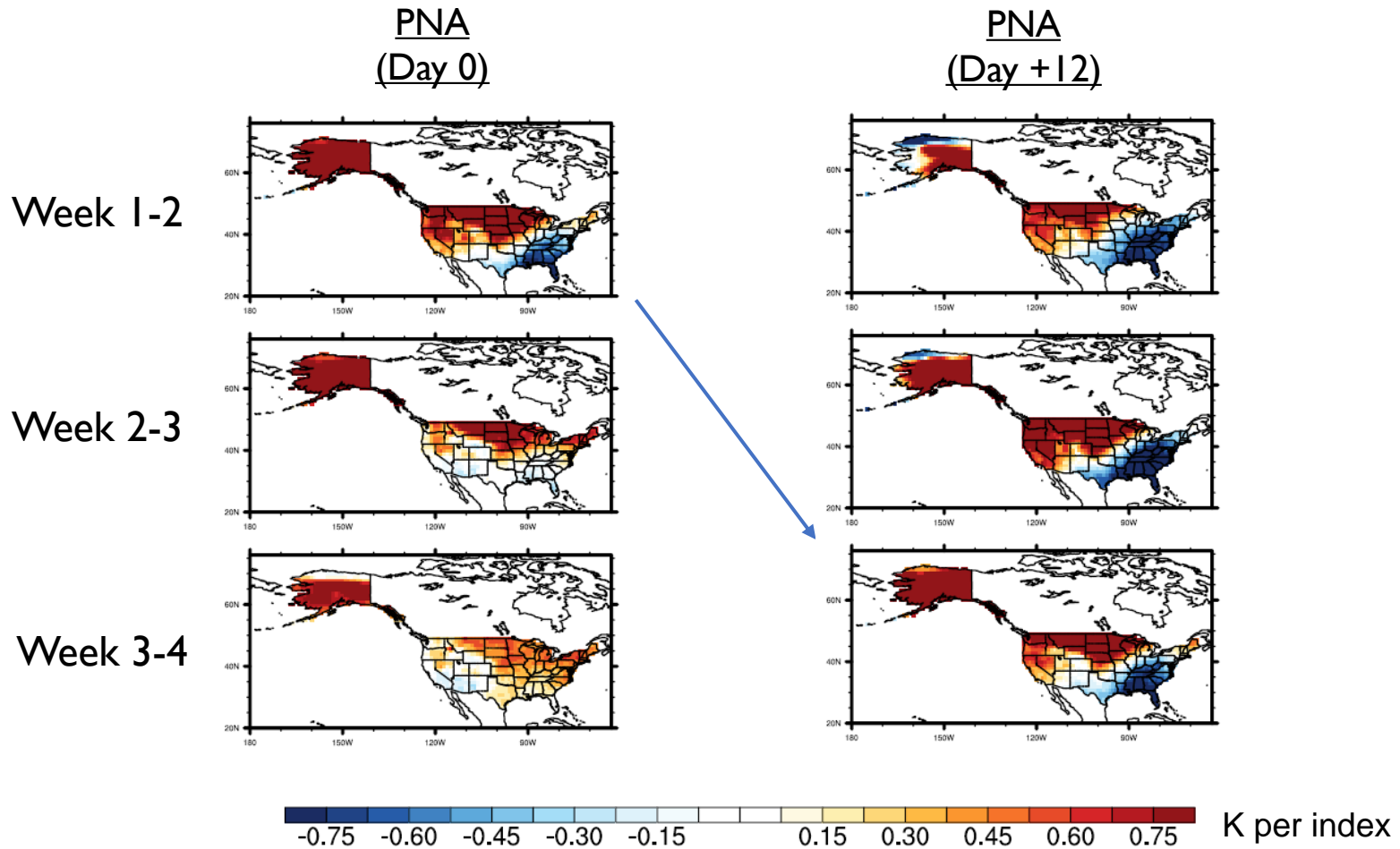
MLR-PNA Schematic (Training)



MLR-PNA Schematic (Verification)

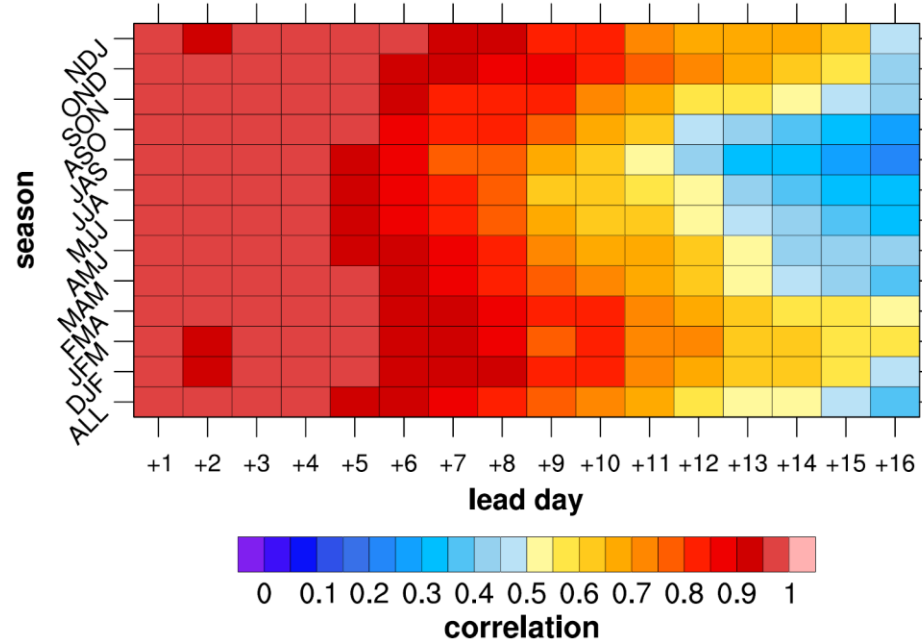


MLR-PNA Regression Coefficients



- Training the MLR-PNA on the observed Day +12 NAO provides a much stronger Week 3-4 signal, so hybridization/bridging the statistical MLR with values of the PNA forecasted by the dynamical models may be ideal.

Observed PNA versus GEFS PNA



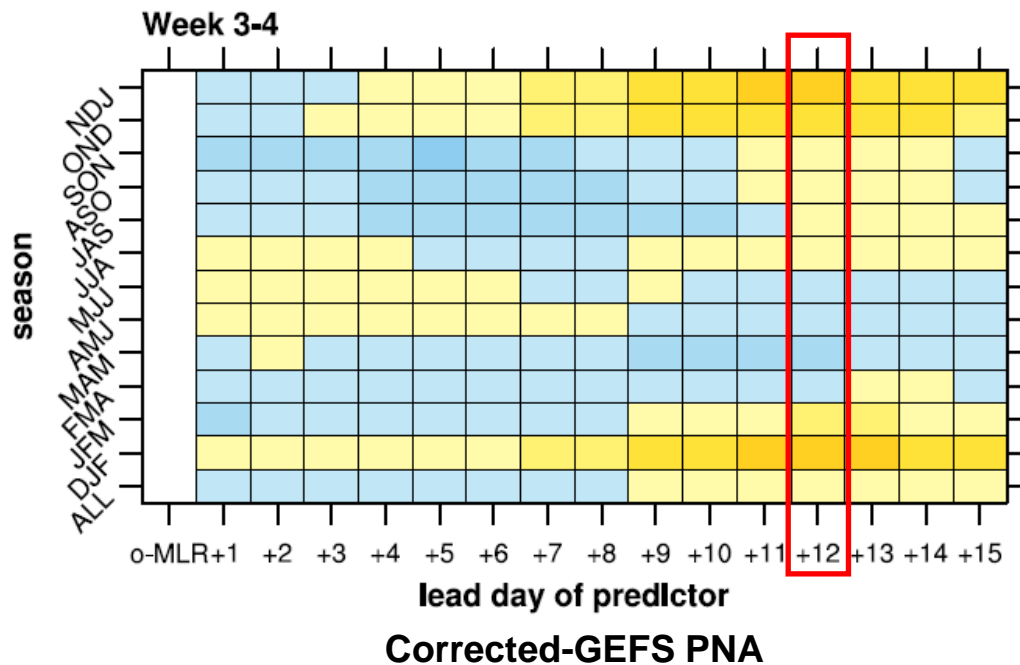
- GEFS forecast versus observations
- The PNA is based on CPC's methods (RPCA on Z500; Barnston & Livezey 1987)

• PNA skill scores peak in winter, with correlations exceeding 0.5 through Day +14, when averaged across all seasons.

original-MLR versus MLR-PNA



Seasonal Skill Score Improvement of the MLR-PNA over the original-MLR



Difference in Week 3-4 TEMPERATURE skill scores

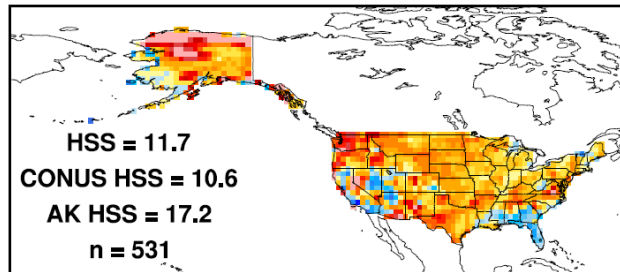
- original-MLR versus MLR-NAO
- **Predictor:** GEFS Days +1 to +15 PNA
- **Verification Period:** 2011-2021, Tuesday and Fridays
- **Key Points:**
 - Generally, the MLR-PNA offers the most improvement during non-summer months.
 - The greatest improvement is provided using the GEFS PNA from Day +12

original-MLR versus MLR-PNA

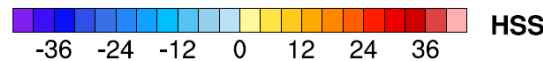
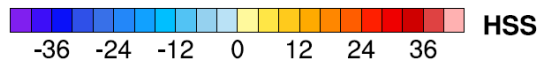
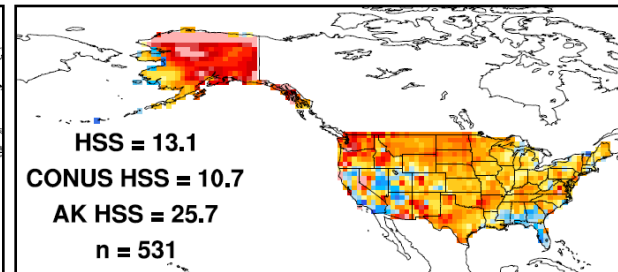
Temperature

original-MLR versus MLR-PNA during NDJFMA (2011-2021)
during All Forecast Initializations

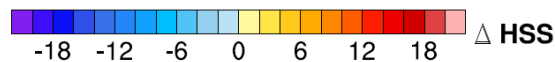
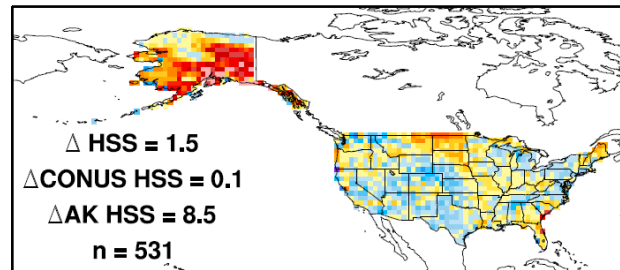
original-MLR



MLR-PNA



MLR-PNA minus original-MLR



Predictor: corrected-GEFS Day +12 PNA (CPC)

Difference in Week 3-4 TEMPERATURE skill scores

- **original-MLR versus MLR-PNA**
- **Predictor:** GEFS Day +12 PNA
- **Verification Period:** 2011-2021, Tuesday and Fridays
- **Additional Conditions:** during November-April only
- **Key Points:**
 - Overall skill scores improve by ~11%.
 - Generally, the MLR-PNA offers improvements over Alaska and does little for CONUS.
 - Skill scores over Alaska improve by ~49%.

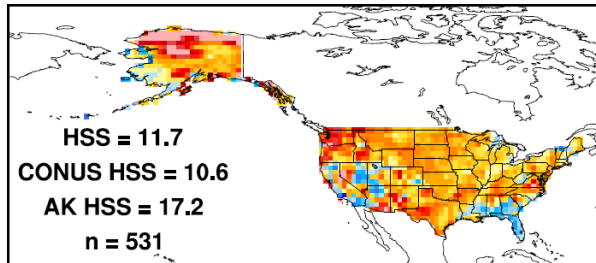
Should we create a merged-MLR,
where we use the MLR-NAO to
forecast for CONUS and the MLR-
PNA to forecast for Alaska?

Original-MLR versus Merged-MLR

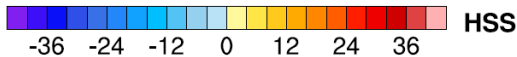
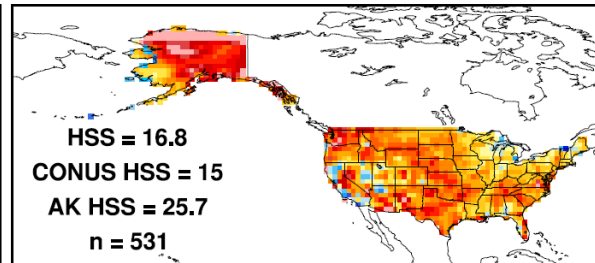
Temperature

original-MLR versus merged-MLR during NDJFMA (2011-2021)
during All Forecast Initializations

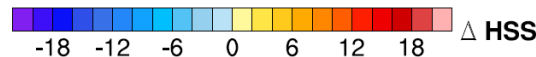
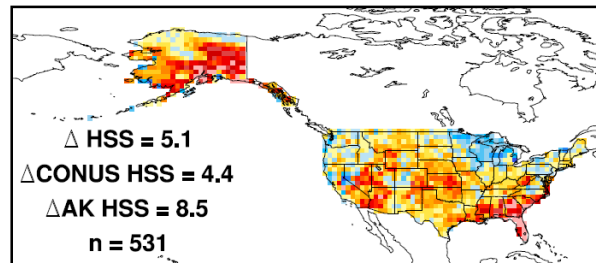
original-MLR



merged-MLR



merged-MLR minus original-MLR



Predictor: corrected-GEFS D+12 PNA (CPC) / D+14 NAO (CPC)

Difference in Week 3-4 TEMPERATURE skill scores

- original-MLR versus merged-MLR
- **Predictor:** GEFS Day +14 NAO & GEFS Day +12 PNA
- **Verification Period:** 2011-2021, Tuesday and Fridays
- **Additional Conditions:** during November-April only
- **Key Points:**
 - Overall skill scores improve by ~44%.
 - Skill score improvements exist nearly everywhere except for the Upper Midwest.

How does the merged-MLR
perform during Forecasts of
Opportunity?

Original-MLR versus Merged-MLR

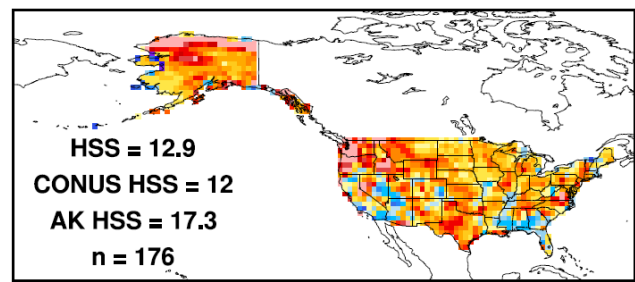


Temperature

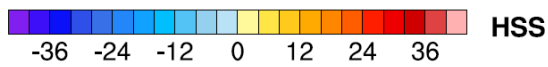
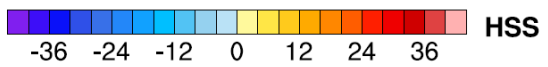
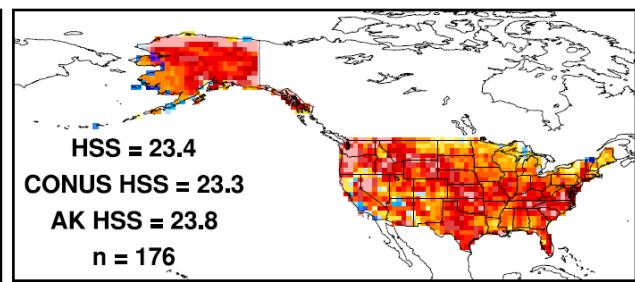
original-MLR versus merged-MLR during NDJFMA (2011-2021)

during Forecast Initializations when $|\text{Day}+0 \text{ NAO}| \geq 0.85$ and $|\text{Day}+0 \text{ NAO}| \geq 0.85$

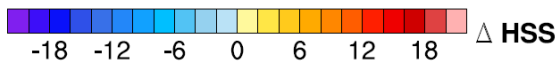
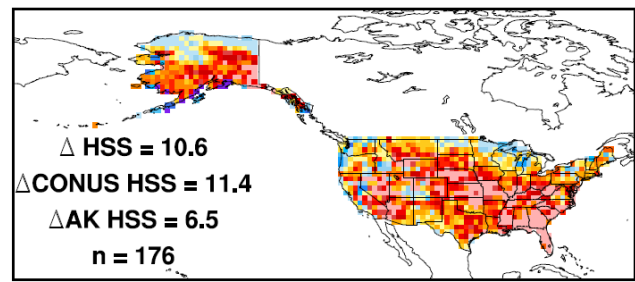
original-MLR



merged-MLR



merged-MLR minus original-MLR



Predictor: corrected-GEFS D+12 PNA (CPC) / D+14 NAO (CPC)

Difference in Week 3-4 TEMPERATURE skill scores

- original-MLR versus merged-MLR
- Predictor: GEFS Day +14 NAO & GEFS Day +12 PNA
- Verification Period: 2011-2021, Tuesday and Fridays
- Additional Conditions: during November-April only & when the observed NAO is amplified on Day 0
- Key Points:
 - Overall skill scores improve by ~82% over CONUS/AK
 - The greatest improvements are located over the Central Plains and Southeast.
 - Forecasts of Opportunity! (Mariotti et al. 2020)

How does the merged-MLR
perform compared to the
GEFSv12?

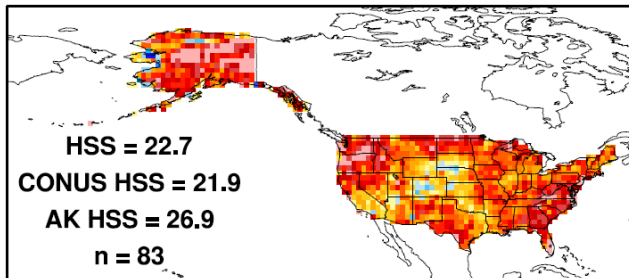
Merged-MLR versus GEFS

Temperature

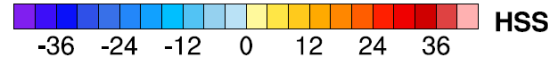
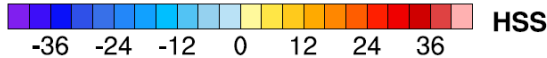
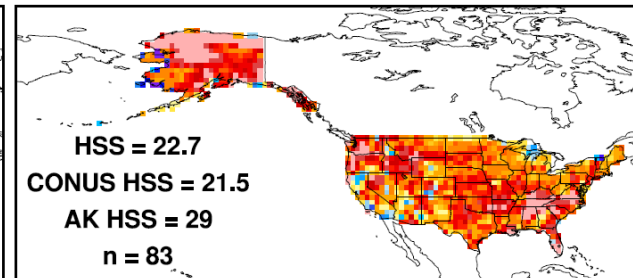
merged-MLR versus GEFSv12 during NDJFMA (2011-2019)

during Forecast Initializations when $|\text{Day}+0 \text{ NAO}| \geq 0.85$ and $|\text{Day}+0 \text{ NAO}| \geq 0.85$

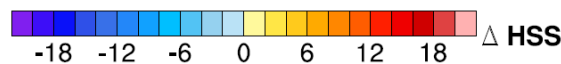
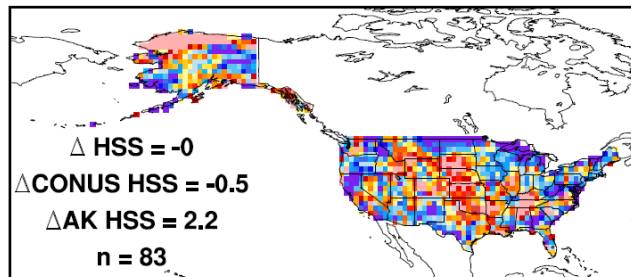
GEFSv12



merged-MLR



merged-MLR minus GEFSv12



Difference in Week 3-4 TEMPERATURE skill scores

- merged-MLR versus GEFS
- Predictor:** GEFS Day +14 NAO & GEFS Day +12 PNA
- Verification Period:** 2011-2019, Thursdays
- Additional Conditions:** during November-April only & when the observed NAO is amplified on Day 0
- Key Points:**
 - The GEFS and merged-MLR perform equally well when the NAO is amplified.
 - Skill score improvements are greatest over northern Alaska and central CONUS.
 - Statistical models have a place in forecasting!

- By using blocking-related predictors, such as the NAO and PNA, we can improve our Week 3-4 statistical models.
- Further, by hybridizing the statistical models with indices forecasted by the dynamical models, we gain the most improvement.
- Finally, this improvement largely occurs during so-called “forecasts of opportunity” when the relative indices are amplified. In such instances, the statistical model performs on-par with the GEFSv12.
- Unfortunately, positive results for precipitation have been elusive, but a few more tests are ongoing.
- Moving forward, we will be experimentally monitoring the merged-MLR’s performance in real-time, with the particular hope that it can provide insight into upcoming episodes of cold during winter.

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Extra Slides



Introduction

- Why do we care about blocking?
- A brief overview of CPC's forecasting process
- What is CPC's Multiple Linear Regression Model (original-MLR)?

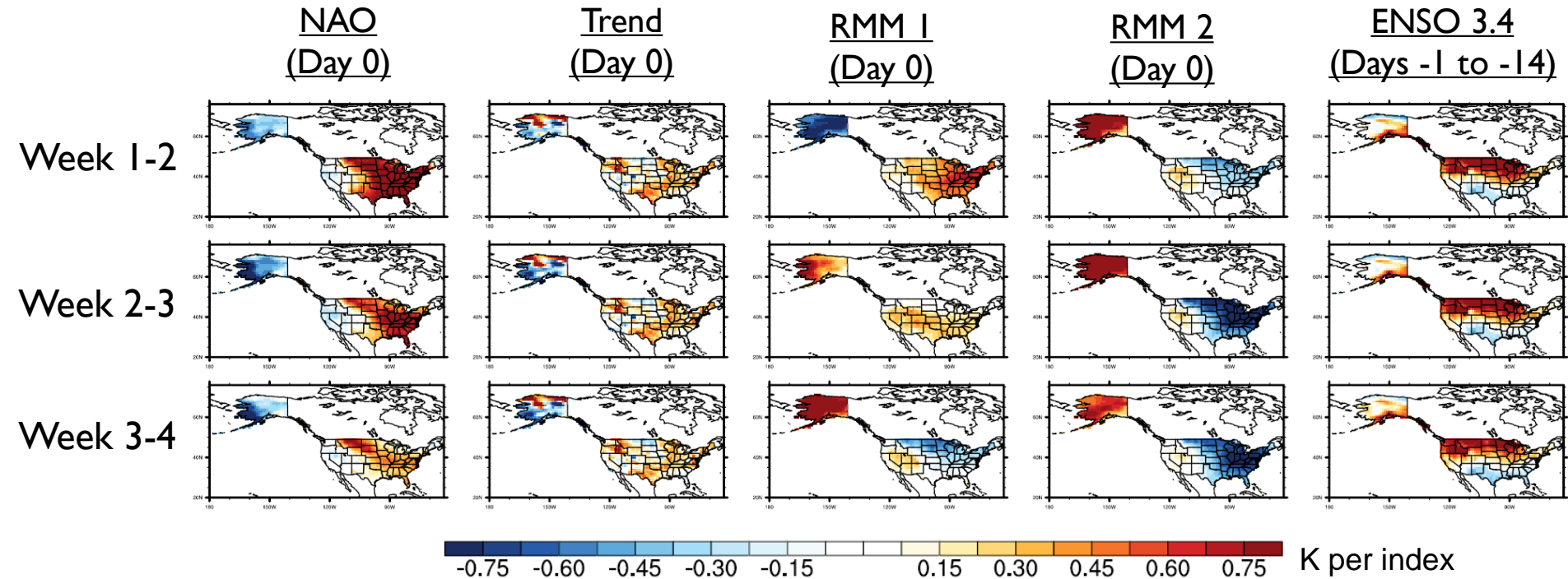
Results

- How does an MLR-NAO perform compared to the original-MLR?
- How does an MLR-PNA perform compared to the original-MLR?
- Should we create a merged-MLR, where we use the MLR-NAO to forecast for CONUS and the MLR-PNA to forecast for Alaska?
- How does the merged-MLR perform during Forecasts of Opportunity?
- How does the merged-MLR perform compared to the GFSv12?

Conclusions

MLR-NAO Regression Coefficients

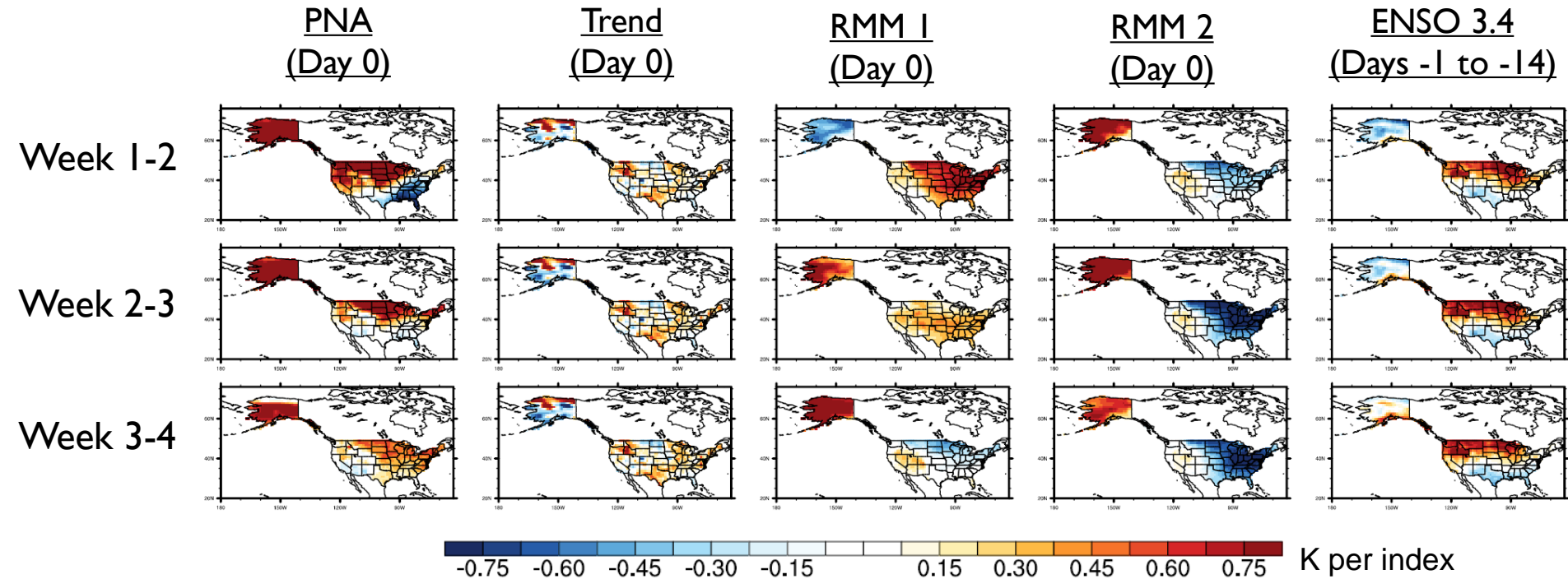
Temperature



- Trend remains constant with lead-time.
- MJO and ENSO signals are large across all leads and persist out to Week 3-4.
- The large NAO signal over CONUS fades significantly by Week 3-4.

MLR-PNA Regression Coefficients

Temperature



- Trend remains constant with lead-time
- MJO and ENSO signals are large across all leads and persist out to Week 3-4
- The PNA signal tends to fade toward Week 3-4